

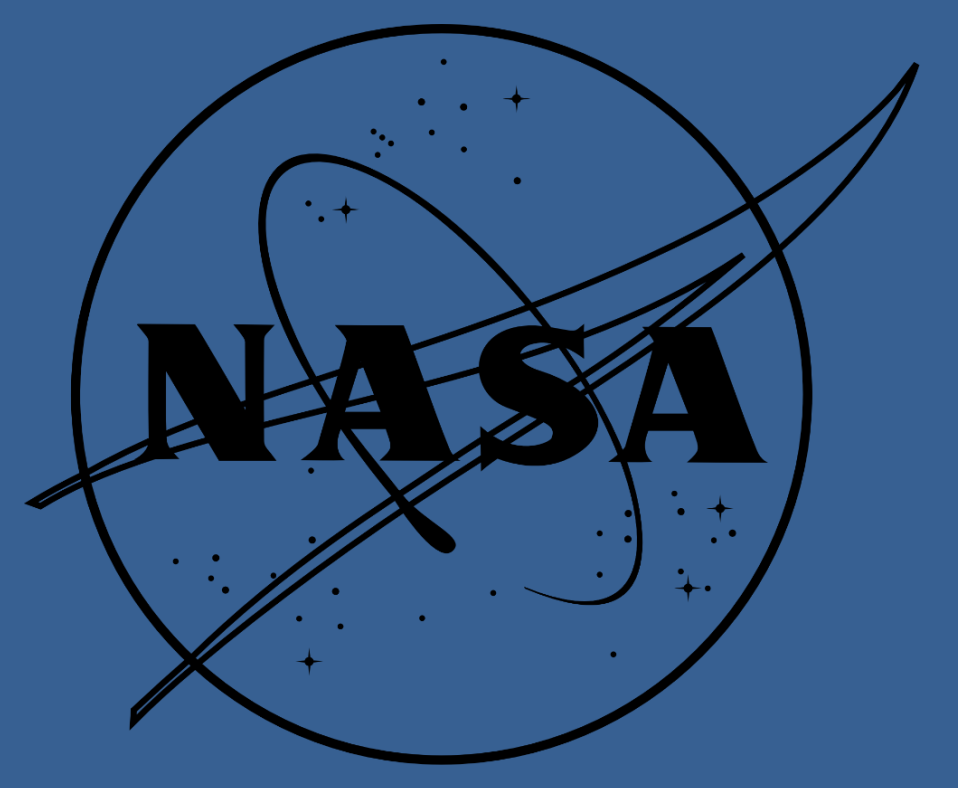


A Word from Alpine Tundra: Watch Out, Forests Are Invading!

Spatio-temporal Dynamics of Alpine Treeline Ecotones in the Western United States

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Introduction

Where do trees survive in mountains? 1) Growing season ≥ 3 months/94 days. 2) Average air temperature during the growing season $\geq 6.4^\circ\text{C}^{2,5}$.

Alpine Treeline Ecotone (ATE) is: an unique and important **ecological transition zone** from subalpine **forests** to alpine **tundra** in global mountain environments (see the red rectangle in Fig. 1).

Why do we care about the ATE? 1) It's an **essential habitat** for numerous species (see Fig. 2). 2) It's relevant to many **ecological functions** such as carbon sequestration, nutrient and water cycling, snow retention, albedo and surface roughness, etc. 3) It is both a potential **at-risk area** and a powerful **indicator** of climate change⁴ (see the red rectangles in Fig. 3).

ATE Characteristics

The **study area** (see the light gray area in Fig. 4) was determined based on a **climate-based estimate** of treeline elevation dataset^{3,5}. Based on the NDVI and elevation data, we defined three **ATE detection characteristics**⁹:

- 1) Abrupt spatial gradient in NDVI:** NDVI typically decreases from montane to alpine zones (see "E1" in Fig. 5 & "C1" in Table 1).
- 2) Intermediate NDVI:** ATEs are typically characterized by intermediary NDVI values (see "E2" in Fig. 5 & "C2" in Table 1). The Gaussian function was constructed based on the categorized NDVIs of the **103 calibration pixels** (see the orange dots in Fig. 4): $a = 1$, $b = 0.44$, and $c = 0.06$.
- 3) Spatial co-variation of elevation and NDVI:** In ATEs, NDVI typically decreases as elevation increases (see "E3" in Fig. 5 & component "C3" in Table 1).

ATE Detection Index

An **ATE detection index (ATEI)**⁹ ranging from 0 to 1 was constructed by fitting a **binomial logistic regression** model of the three standardized components of the **141 validation pixels** (see the blue diamonds in Fig. 4). The coefficients of C1, C2, and C3 were 0.44, 0.58, and 0.56, respectively. The intercept was -1.47.

ATEI detection **accuracy assessment**⁹: 1) 100-time repeated 10-fold **cross-validation**: average overall accuracy = 0.713 (± 0.111), average Kappa coefficient = 0.426 (± 0.221). 2) An independent **field-based** ATE dataset¹⁰ (see the yellow triangles in Fig. 4): Pearson's $r = 0.98$.

Transect Generation

Annual ATEI calculation from 1984 to 2018 (see Fig. 6) was based on: 1) annual max. NDVIs of **LANDSAT-5**, 7, and 8⁸, and 2) **AW3D30** elevation data^{6,7}.

Random transect generation (see Fig. 7): 1) Divide the study area into **Thiessen polygons**. 2) Extract the **steepest transect** in each polygon. 3) Select transects with an average annual **max. ATEI** over 0.75 and a **length** between 500 m and 3 km. 4) Create a 30-m **buffer** of each selected transect.

Spatio-temporal Dynamics

ATE elevation trend analysis (see Fig. 8): 1) Extract the elevation of the pixel with the **max. ATEI** in each year within each transect. 2) Perform a **robust linear regression** between time and the extracted annual ATE elevation.

Estimated ATE elevation change from 1984 to 2018 in **32,186 transects** (see Fig. 11): 1) **50.9%** moving **uphill** (see the red area in Fig. 10) at **6.25 m/decade**. 2) 22.9% moving downhill at -5.66 m/decade (see the blue area in Fig. 10). 3) 26.2 % no significant change (elevation trend between -1 and 1 m/decade, see the white area in Fig. 10).

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Fig. 1. An example of the Alpine Treeline Ecotone in Glacier National Park, MT.

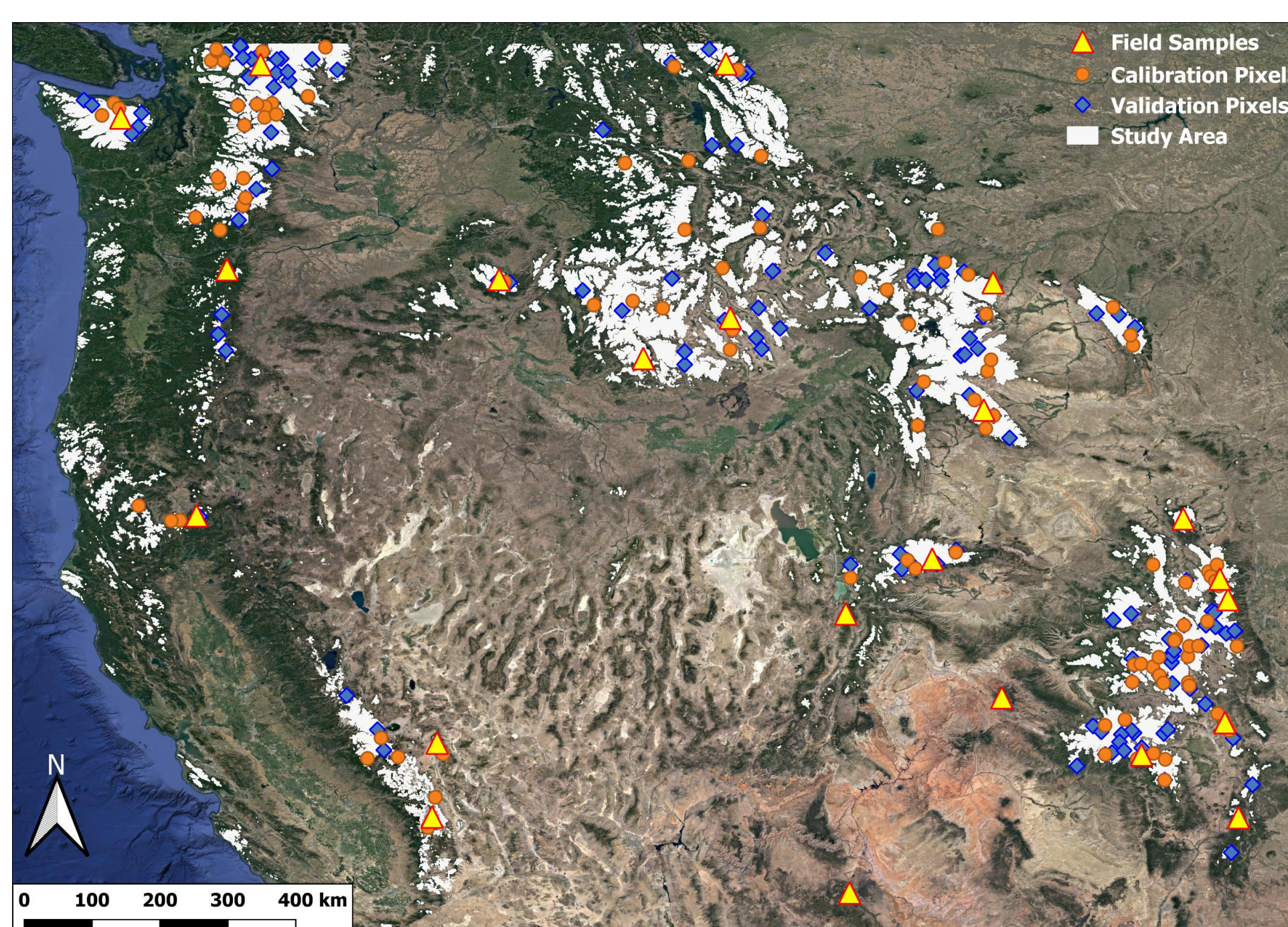


Fig. 4. Geographic information of the study area and calibration/validation pixels in this study⁹ (background image¹). The yellow triangles represent the ground sampling sites located within the study area, which are from a published ATE research¹⁰.

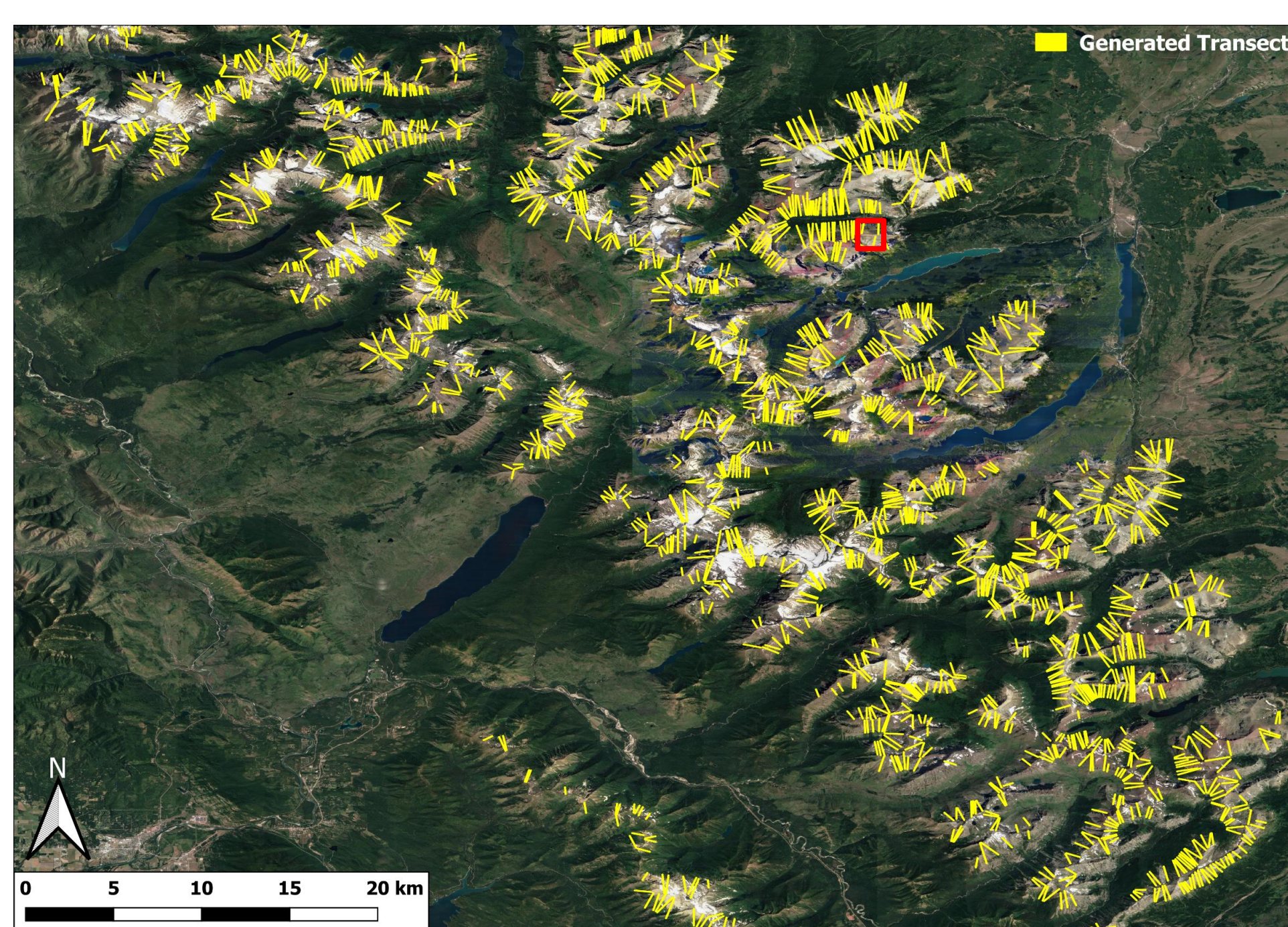


Fig. 7. Selected transects in Glacier National Park, MT (background image¹). The red rectangle shows the location of a transect example.

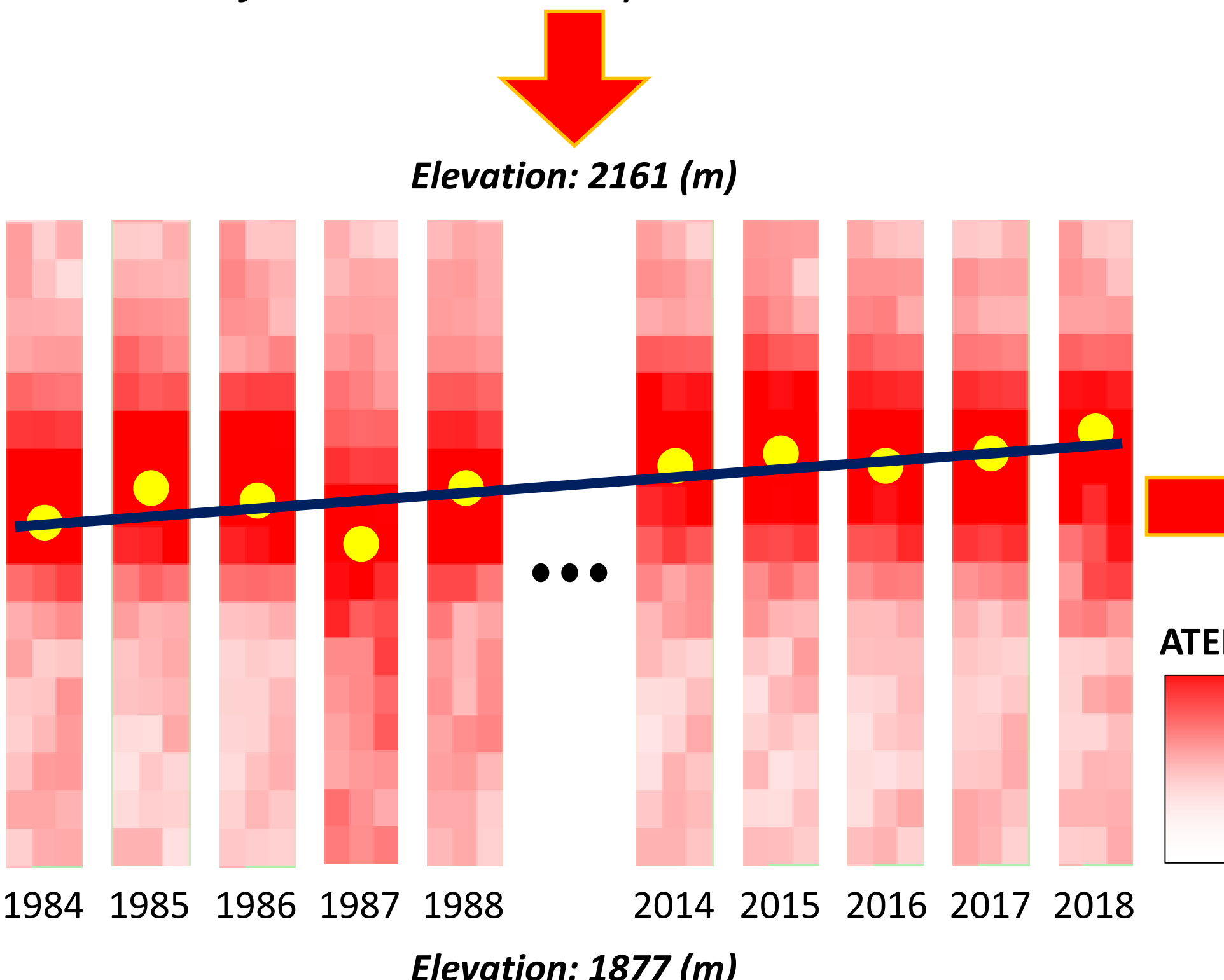


Fig. 8. Annual ATEIs from 1984 to 2018 in a transect example (see the red rectangle in Fig. 7). The elevation of pixels within the transect varies from 1877 m a.s.l. to 2161 m a.s.l. The yellow dots indicate the pixel with the max. ATEI in each year. The dark blue line represents the trend line of the extracted annual ATE elevation.

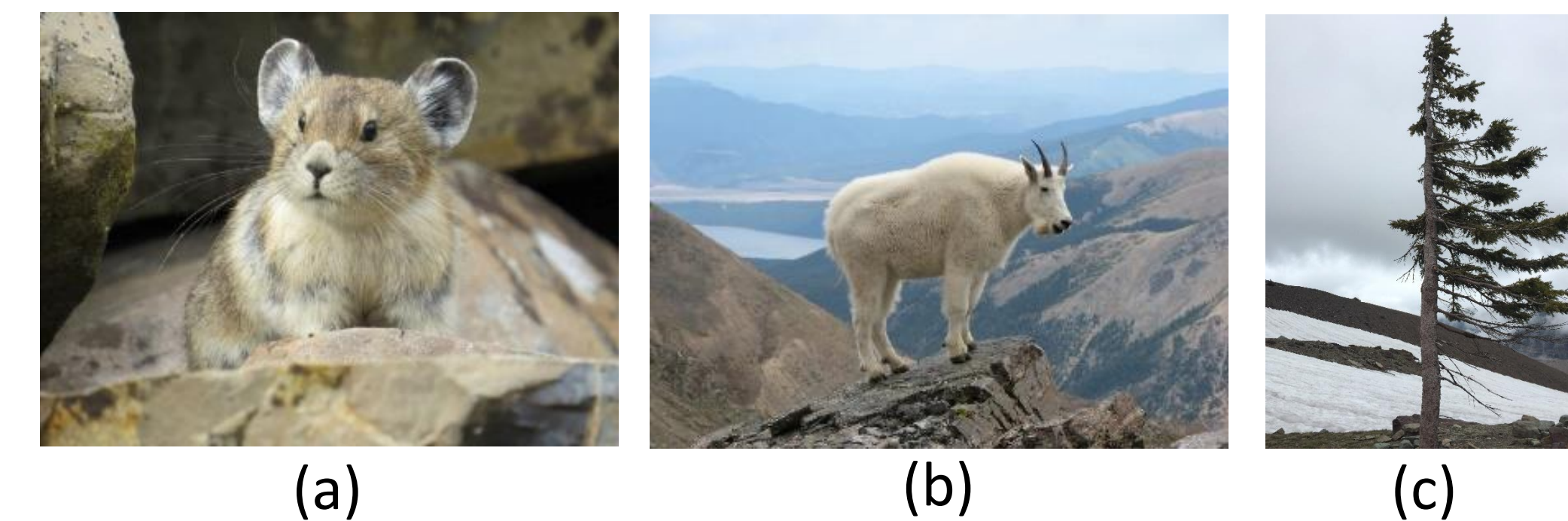


Fig. 2. Several species in ATEs: a) American pika (*Ochotona princeps*, source: Will Thompson photo, courtesy of USGS), b) Mountain goat (*Oreamnos americanus*, source: Darklich14 photo, courtesy of Wikimedia Commons), and c) a flag tree in Glacier National Park, MT.

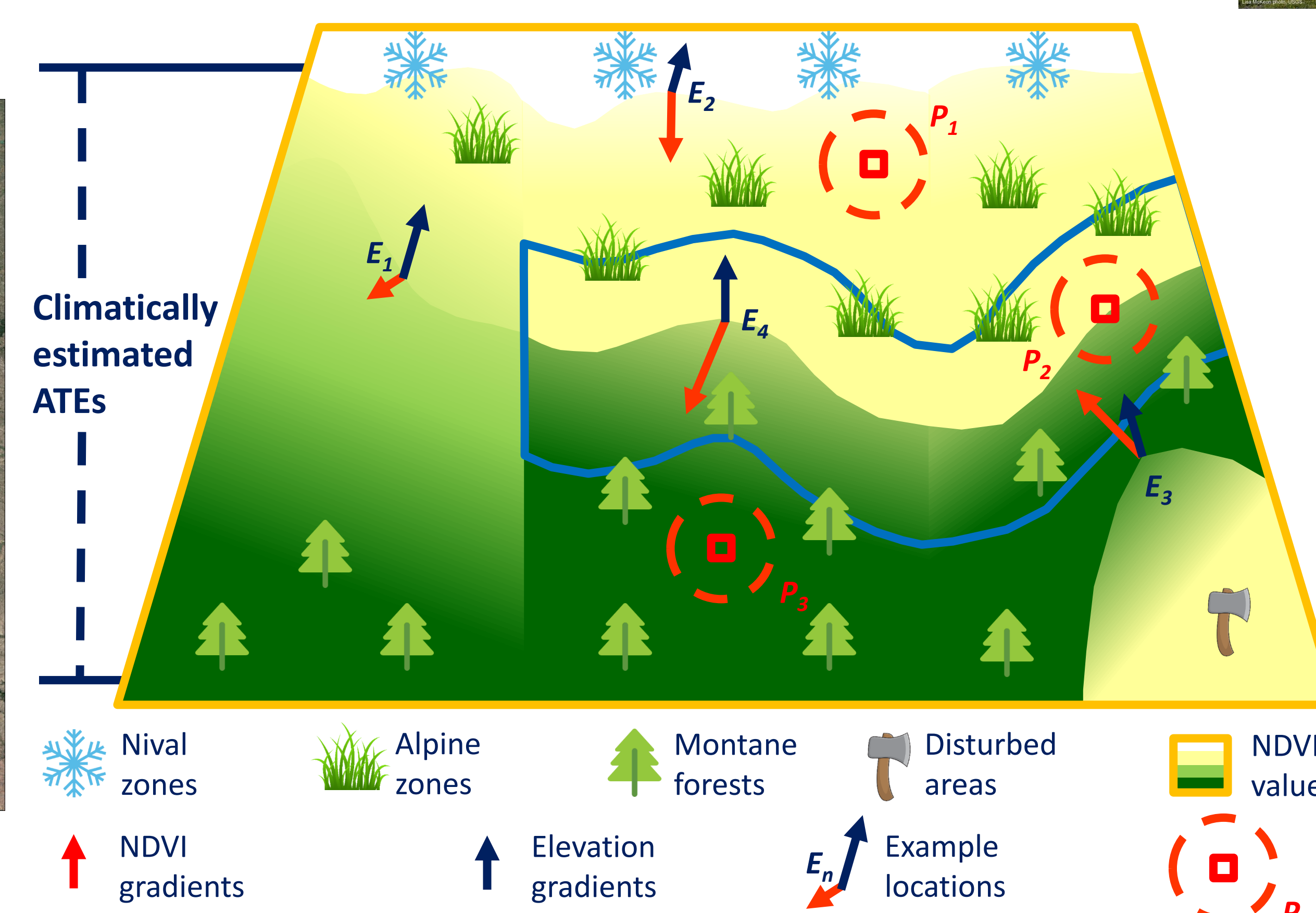


Fig. 5. Spatial detection of ATEs in a simplified example of mountain environment⁹.

$$\ln \frac{ATEI}{1-ATEI} = \beta_0 + \sum_{i=1}^3 \beta_i C_i \quad (i = 1, 2, 3)$$



Fig. 6. Annual ATEIs from 1984 to 2018 in Glacier National Park, MT (background image¹).

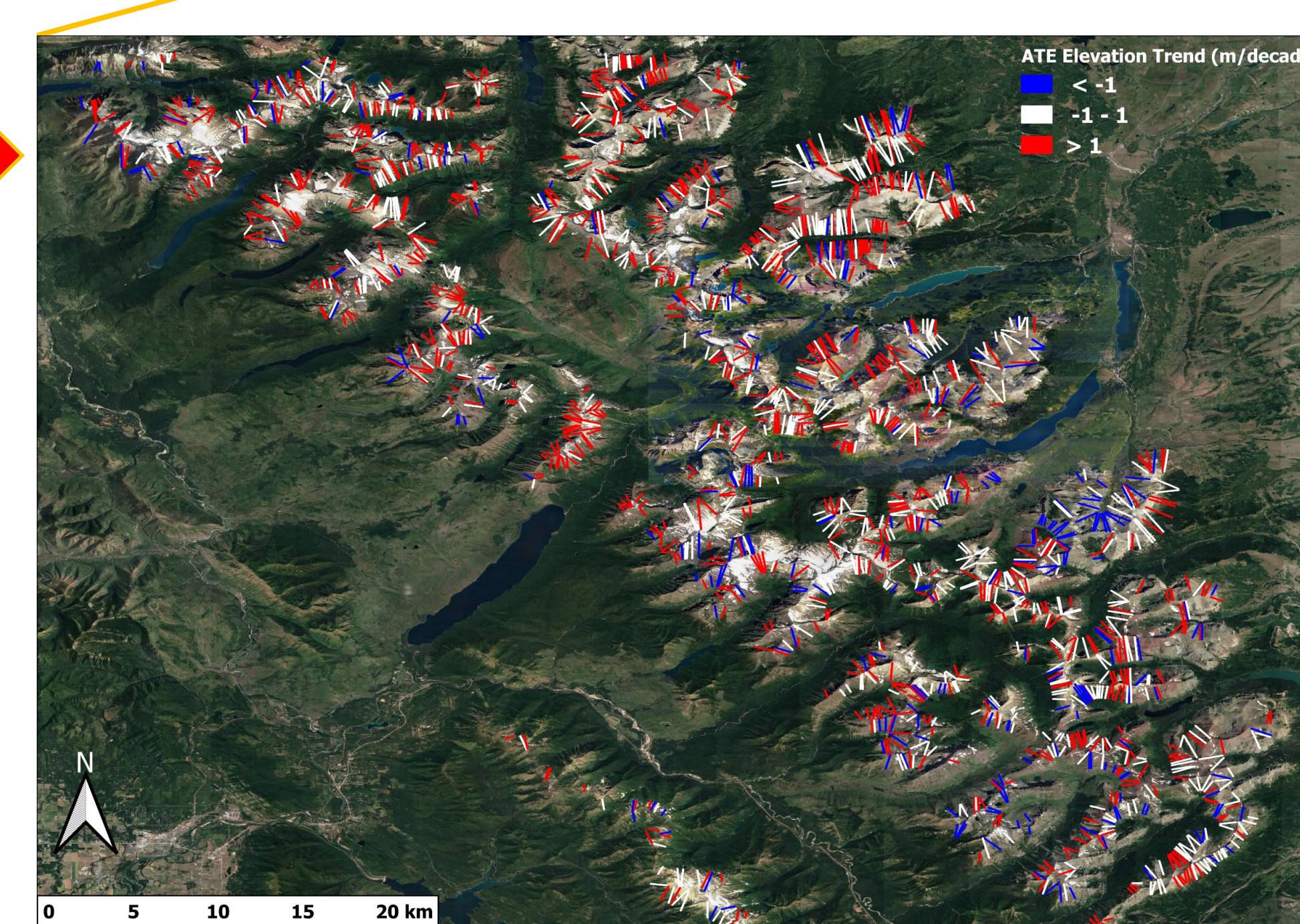


Fig. 9. ATE elevation trends of all selected transects in Glacier National Park (background image¹).

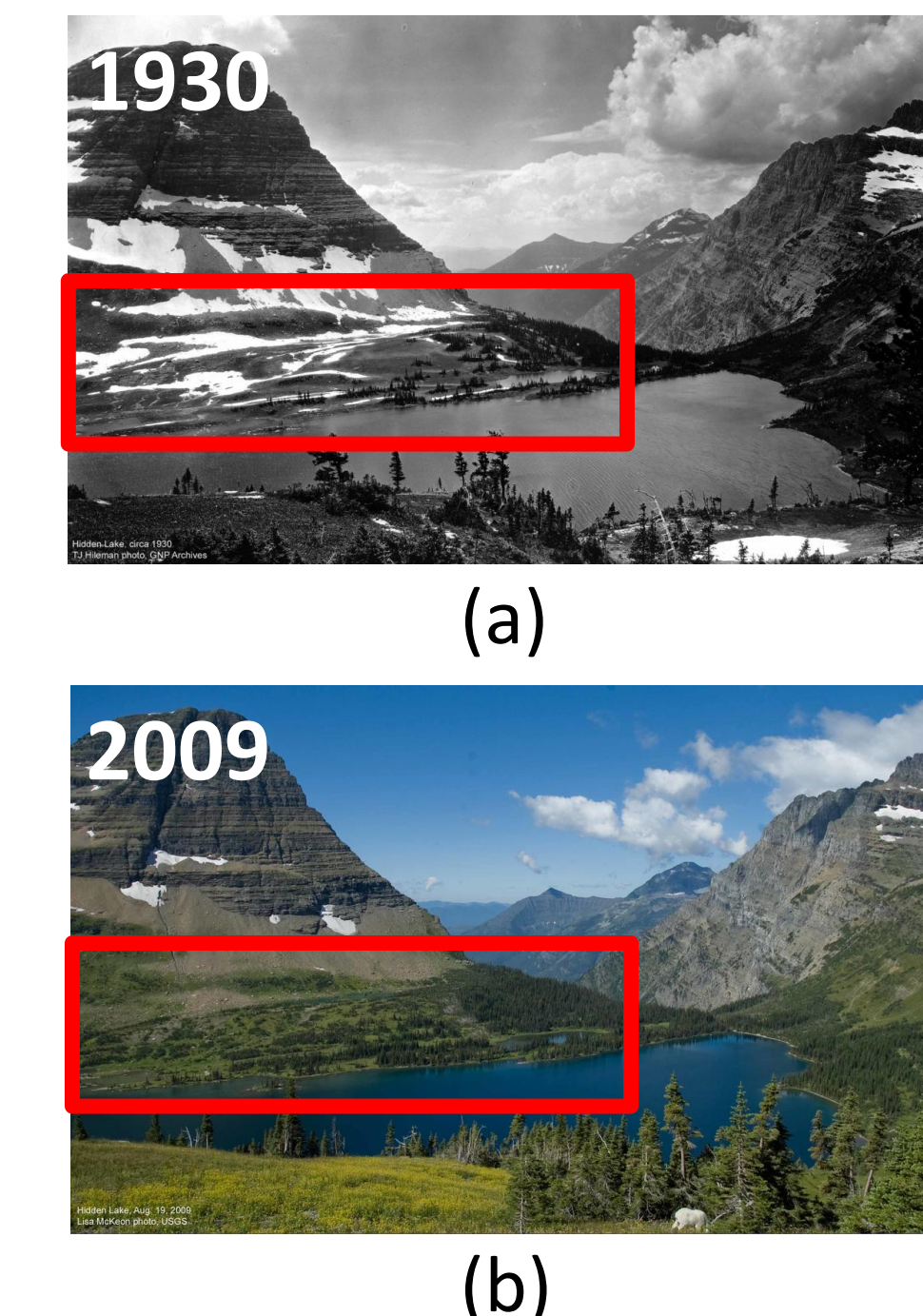


Fig. 3. Hidden Lake in Glacier National Park, MT in: a) 1930 (source: T.J. Hileman photo, courtesy of Glacier National Park Archives) and b) 2009 (source: Lisa McKeon photo, courtesy of USGS).

Table 1 Defined characteristics of ATEs in this study⁹: 1) C_1 quantifies the magnitude of NDVI gradient ∇f_{NDVI} and depresses areas where NDVI varies gradually (see "E1" in Fig. 5). 2) C_2 is a Gaussian function of f_{NDVI} , which depresses areas with very low or very high NDVI values, (see "E2" in Fig. 5). 3) ∇f_{Elev} represents the elevational gradient². C_3 depresses areas where ∇f_{NDVI} and ∇f_{Elev} are in similar directions ($\theta < 90^\circ$ or $\theta > 270^\circ$, see "E3" in Fig. 5). Here, n is set to 10.

No.	Algorithm	Component
C1	$\begin{bmatrix} 0.5 & 0.3 & 0.3 \\ 0.6 & 0.5 & 0.4 \\ 0.7 & 0.7 & 0.6 \end{bmatrix} \nabla f_{NDVI}$	$\ \nabla f_{NDVI}\ $
C2	$\frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(f_{NDVI}-b)^2}{2\sigma^2}\right)$	$\frac{(f_{NDVI}-b)^2}{2\sigma^2}$
C3	$\frac{(1 - \cos \theta)^n}{2^n}$	$\frac{(1 - \cos \theta)^n}{2^n}$

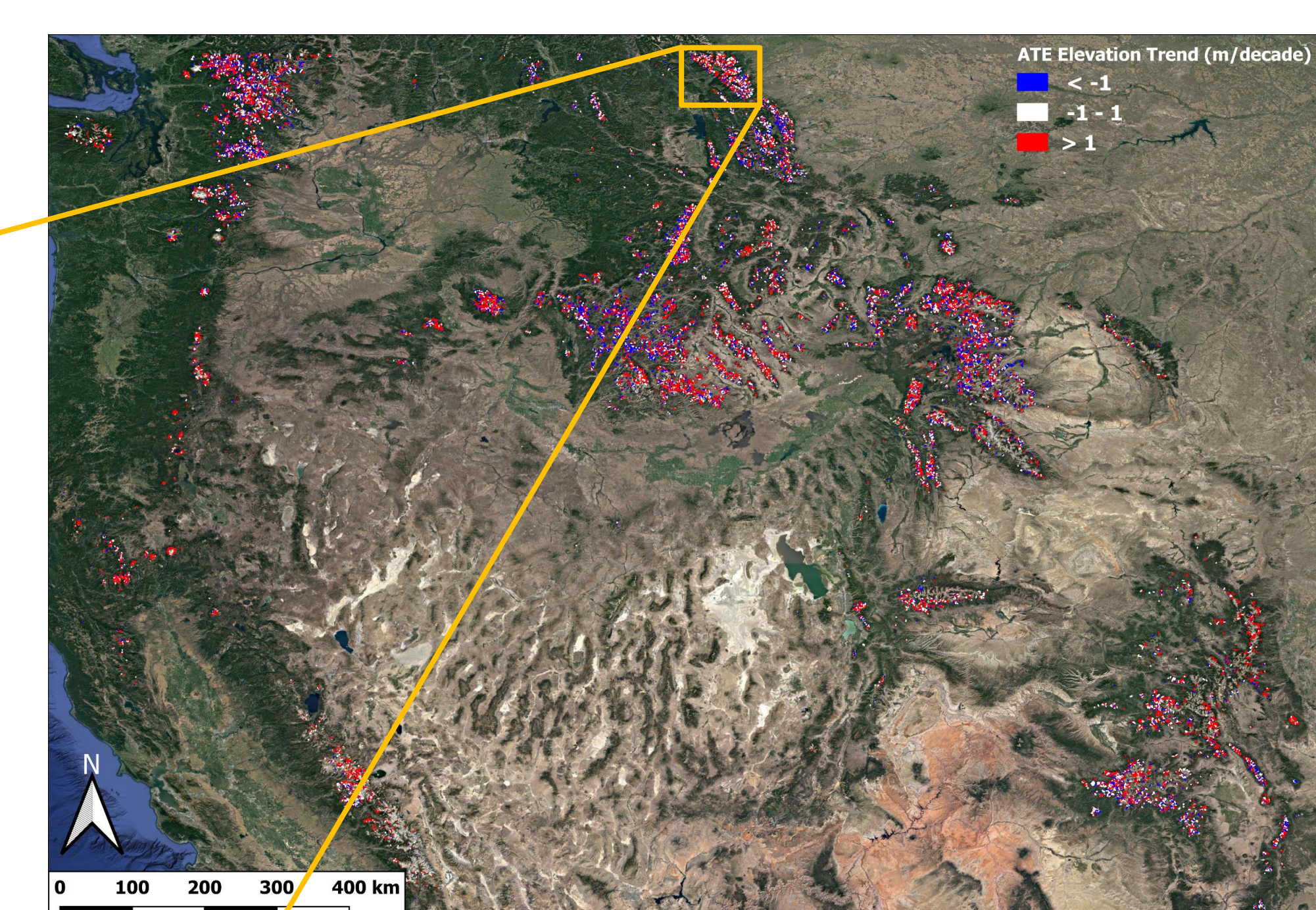


Fig. 10. ATE elevation trends of all selected transects in the study area (background image¹).

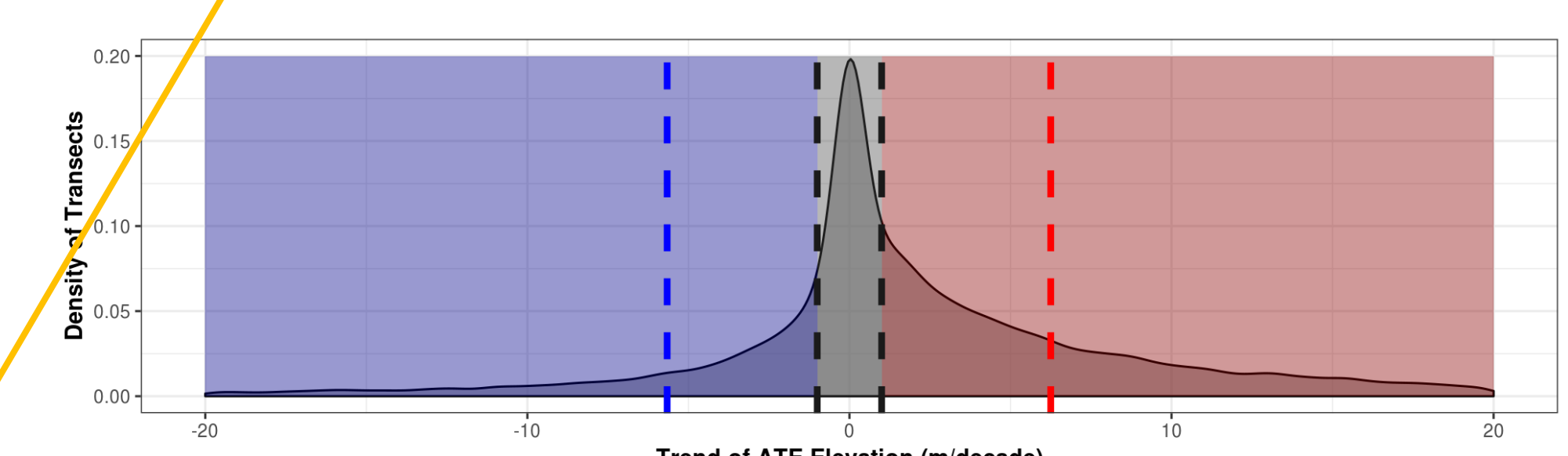


Fig. 11. Distribution of the elevation trends of all selected transects in the study area.

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